

# Neural Formatting for Spreadsheet Tables

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## Abstract

Spreadsheets are popular and widely used for data presentation and management, where users create tables in various structures to organize and present data. Table formatting is an important yet tedious task for better exhibiting table structures and data relationships. However, without the aid of intelligent tools, manual formatting remains a tedious and time-consuming task.

In this paper, we propose CellGAN, a neural formatting model for learning and recommending formats of spreadsheet tables (Dong et al. 2020b). Based on a novel conditional generative adversarial network (cGAN) architecture, CellGAN learns table formatting from real-world spreadsheet tables in a self-supervised fashion without requiring human labeling. In CellGAN we devise two mechanisms, row/column-wise pooling and local refinement network, to address challenges from the spreadsheet domain. We evaluate the effectiveness of CellGAN against real-world datasets using both quantitative metrics and human perception studies. The results indicate remarkable performance gains over rule-based methods, graphical models or direct application of the state-of-the-art cGANs used in visual synthesis tasks. Neural Formatting is the first step towards auto-formatting for spreadsheet tables with promising results.

In this paper, we propose techniques for table formatting style transfer, i.e., to automatically format a target table according to the style of a reference table (Dong et al. 2020a). Considering the latent many-to-many mappings between table structures and formats, we propose CellNet, which is a novel end-to-end, multi-task model leveraging conditional Generative Adversarial Networks (cGANs) with three key components to (1) model and recognize table structures; (2) encode formatting styles; (3) learn and apply the latent mapping based on recognized table structure and encoded style, respectively. Moreover, we build up a spreadsheet table corpus containing 5,226 tables with high-quality formats and 784 tables with human-labeled structures. Our evaluation shows that CellNet is highly effective according to both quantitative metrics and human perception studies by comparing with heuristic-based and other learning-based methods.

	A	B	C	D	E	F	G
1	CHARACTERISTIC	Employed wage and salary workers		Non employed usual weekly earnings			
2							
3		Total	Percent	Total	Number		
4							
5			Union	Not		Not	
6			members	union		represented	
7				members		by union	
8	Total	129378	0.124	0.876	8960	4904	5056
9	Age						
10	16 to 34 years old	47981	0.157	0.843	3421	1342	2079
11	35 to 54 years old	59495	0.294	0.706	3427	1867	1560
12	55 years and over	21901	0.256	0.744	3112	1695	1417
13	Industry						
14	Private sector	108073	0.076	0.924	7429	3238	4191
15	Agriculture	1057 NA	NA	NA	NA	NA	NA
16	Nonagriculture	107016 NA	NA	NA	NA	NA	NA
17	Public sector	21305	0.0368	0.9632	1531	1666	865

(a) A table with default formats

	A	B	C	D	E	F	G
1	CHARACTERISTIC	Employed wage and salary workers		Not employed usual weekly earnings			
2							
3			Percent			Number	
4		Total	Union	Not	Total	Represented	Not
5			members	union		by union	represented
6				members			by union
7	Total	129,378	12.4%	87.6%	9,960	4,904	5,056
8	Age						
9	16 to 34 years old	47,981	15.7%	84.3%	3,421	1,342	2,079
10	35 to 54 years old	59,495	29.4%	70.6%	3,427	1,867	1,560
11	55 years and over	21,901	25.6%	74.4%	3,112	1,695	1,417
12	Industry						
13	Private sector	108,073	7.6%	92.4%	7,429	3,238	4,191
14	Agriculture	1,057	NA	NA	NA	NA	NA
15	Nonagriculture	107,016	NA	NA	NA	NA	NA
16	Public sector	21,305	3.7%	96.3%	2,531	1,666	865

(b) A table with human-crafted formats

Figure 1: Comparison of an example spreadsheet table with default formats and human-crafted formats.

## References

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